

## Recall: Homotopy Continuation

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- Form the **homotopy** system  $H(x, t) = (1 - t) \cdot F(x) + t \cdot G(x)$
- Use predictor-corrector methods to **track** the numerical solutions as  $t$  moves from  $t = 1$  to  $t = 0$ .

# Homotopy Tracking

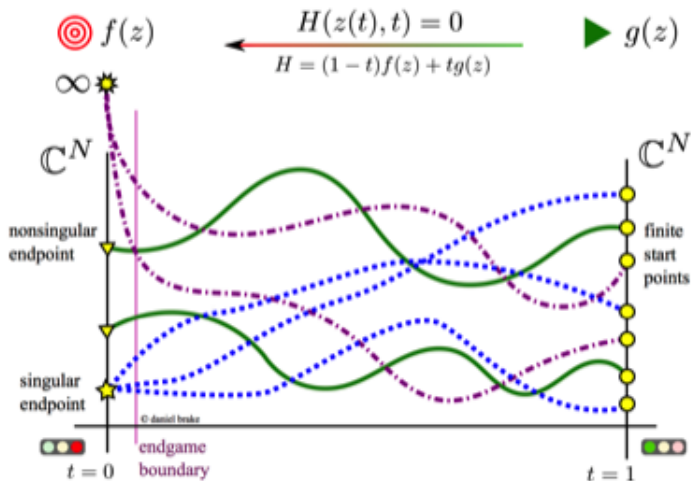


Figure: Homotopy Continuation Illustration (Dani Brake)

## Theorem (Likelihood Geometry Group)

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Let  $\hat{\theta}_{win}$  and  $\hat{\theta}_{stat}$  be the respective MLEs and let  $\gamma$  denote the path of the homotopy whose start point (at  $t = 1$ ) corresponds to  $\hat{\theta}_{win}$ . Then, the endpoint of  $\gamma$  (at  $t = 0$ ) is  $\hat{\theta}_{stat}$ .

- By *Birch's Theorem*, a homotopy between the two systems is given by

$$H(\theta, t) := t \left( A\hat{\rho}_{stat} - \frac{1}{N}Au \right) + (1 - t) \left( A\hat{\rho}_{win} - \frac{1}{N}Au \right)$$

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- Left to show tracking paths do not intersect (we show the Jacobian matrix of the system has always full rank)



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- Knowing how scaling vectors  $c$  affect the ML degree of a particular toric model  $V_A$  allows us to find a convenient  $c_{win}$  (e.g. such that the model has *low* ML degree).
- By the Theorem, we can now find the MLE  $\hat{\theta}_{win}$  and track its unique homotopy path to find the original MLE of interest  $\hat{\theta}_{stat}$ .



## Example (Veronese revisited)

Recall

$$A = \begin{bmatrix} 0 & 1 & 2 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 & 1 & 2 \end{bmatrix},$$

with  $u = (1, 3, 5, 7, 9, 2)$ . Here  $c_{stat} = (1, 1, 1, 1, 1, 1)$ .

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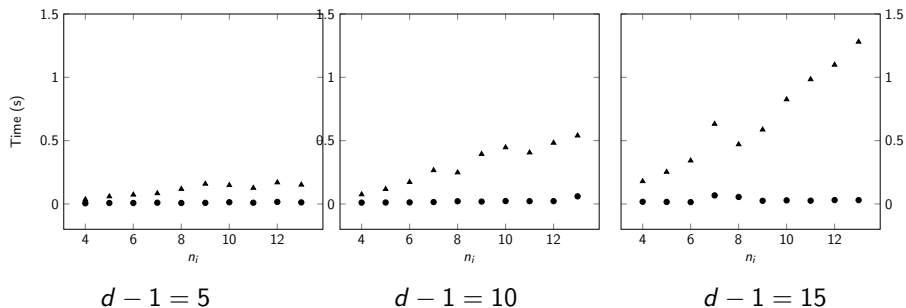
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# Success story



**Figure:** Running times of iterative proportional scaling (triangles) versus path tracking (circles) on rational normal scrolls. Average of 7 trials.